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### BRL

A SMALL SAMPLE POWER STUDY
OF THE ANDERSON-DARLING STATISTIC
AND A COMPARISON WITH THE KOLMOGOROV
AND THE CRAMÉR-VON MISES STATISTICS

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**DECEMBER 1989** 



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### I. INTRODUCTION

Consider a random sample  $X_1, X_2, ..., X_n$  from a population with a continuous distribution function. One method of testing the hypothesis that the n observations come from a population with a specified distribution function F(x) is by a chi-square test. This test requires a subjective partitioning of the real line R and a comparision of the empirical histogram with the hypothetical histogram. A more objective method, is to compare the empirical distribution function  $F_n(x)$  with the hypothetical distribution function F(x). The empirical distribution function based on n observations is defined as  $F_n(x) = \frac{k}{n}$  if exactly k observations are less than or equal to x for k = 0, 1, ..., n.

To compare the empirical and hypothetical distribution functions a measure of their difference is required. Addressing this, Anderson and Darling [1952] considered the following metrics in function space:

$$W_n^2 = n \int_{-\infty}^{\infty} [F_n(x) - F(x)]^2 \psi[F(x)] dF(x)$$
 (1)

and

$$K_{n} = \sup_{-\infty < x < \infty} \sqrt{n} |F_{n}(x) - F(x)| \sqrt{\psi[F(x)]}.$$
 (2)

Samples producing large values of  $W_n^2$  (or  $K_n$ ) lead to rejection of the null hypothesis that the population distribution function is F(x). One of the contributions of Anderson and Darling was the incorporation of a non-negative weight function  $\psi$  in (1) and (2). By a suitable choice for  $\psi$ , specific ranges of values of the random variable X, corresponding to different regions of the distribution F(x), may be emphasized. For  $\psi[F(x)] \equiv 1$ ,  $W_n^2$  becomes the Cramérvon Mises statistic [Cramér, 1928 and von Mises, 1931] and  $K_n$  becomes the Kolmogorov statistic [Kolmogorov, 1933].

The tails of the distribution function will be accentuated in the investigation detailed in this paper; Anderson and Darling suggest using

$$\psi[F(x)] = \frac{1}{F(x)[1 - F(x)]}$$
.

With this choice for the weighting function, metric (1) becomes the basis for the Anderson-Darling statistic.

In Section II, the Anderson-Darling test statistic is developed; in Section III, the most accurate tabulation to date of the test statistic is provided. In Section IV, the description and the results of a power study are given in which the Anderson-Darling, the Cramér-von Mises, and the Kolmogorov statistics are compared.

### II. THE ANDERSON-DARLING STATISTIC

For a fixed value of the random variable X, say X = x, the empirical distribution function  $F_n(x)$  is a statistic, since it is a function of the sample values  $x_1, x_2, ..., x_n$ . The distribution of this statistic is established as a lemma.

**Lemma (1):** If  $F_n(x)$  is the empirical distribution function corresponding to a random sample  $X_1, X_2, ..., X_n$  of size n from a distribution  $H(\cdot)$ , then for a fixed x,  $nF_n(x)$  is distributed binomial (H(x), n).

Proof:

$$P(nF_n(x) = k) = P(\text{exactly } k \text{ values } x_i \le x), \text{ for } k = 0, 1, ..., n.$$

Let  $Z_i = I_{(-\infty, x]}(X_i)$ , where the indicator function I is defined as

$$I_{(-\infty, x]}(X_i) = \begin{cases} 1, & \text{if } -\infty < X_i \le x \\ 0, & \text{otherwise} \end{cases}$$

Then  $\Sigma Z_i$  counts the number of sample values  $x_i \leq x$ . Here each  $Z_i \sim \text{Bernoulli}(H(x))$ , so  $\Sigma Z_i \sim \text{binomial}(H(x), n)$ . Therefore,

$$\begin{split} P(nF_n(x) = k) &= P(\text{exactly } k \text{ values } x_i \le x) \\ &= P(\Sigma Z_i = k) \end{split}$$

$$&= \binom{n}{k} H(x)^k \left(1 - H(x)\right)^{n-k} . \blacksquare$$

From Lemma 1,

$$E[F_n(x)] = \frac{1}{n} E[nF_n(x)] = H(x)$$

and

$$Var[F_n(x)] = \frac{1}{n^2} Var[nF_n(x)] = \frac{1}{n} H(x) [1 - H(x)].$$
 (3)

To assist in the determination of a suitable weighting function  $\psi(\cdot)$ , that is, a function that will weight more heavily values in the tails of the distribution F(x) at the expense of values closer to the median, consider the expectation of the squared discrepancy  $E\left[F_n(x)-F(x)\right]^2$ . It is important to keep in mind that the value x is fixed, so F(x) is a constant, and the expectation is with respect to the random variable  $F_n(x)$  whose distribution was established in Lemma 1. Then

$$n E [F_n(x) - F(x)]^2 = n E [F_n(x) - H(x) + H(x) - F(x)]^2$$

$$= n E [\{F_n(x) - H(x)\} - \{F(x) - H(x)\}]^2$$

which, after algebraic manipulation (Appendix A) yields the variance and bias<sup>2</sup>

$$= n \left[ \frac{1}{n} \left\{ H(x) \{ 1 - H(x) \} \right\} + \left\{ F(x) - H(x) \right\}^2 \right]. \tag{4}$$

Under the null hypothesis  $H_0$ :  $H(x) = F(x) \quad \forall x, (4)$  becomes

n E 
$$\left[F_n(x) - F(x)\right]^2 = F(x)[1 - F(x)]$$
. (5)

Anderson-Darling chose as a weighting function,  $\psi[F(x)] = \frac{1}{F(x) [1 - F(x)]}$ .

Weighting by the reciprocal of (5) takes into consideration the variance of the statistic  $F_n(x)$  and also maintains the objective of accentuating values in the tails of F(x).

With this choice of weighting function and without loss of generality assuming  $x_1 \le x_2 \le ... \le x_n$ , let F(x) = u, dF(x) = du, and  $F(x_i) = u_i$ . Then the Anderson-Darling test statistic (6) can be rewritten as expression (7) by expansion and integration (Appendix B).

$$W_n^2 = n \int_{-\infty}^{\infty} \frac{[F_n(x) - F(x)]^2}{F(x) [1 - F(x)]} dF(x) , \qquad (6)$$

$$W_n^2 = -n - \frac{1}{n} \sum_{i=1}^{n} \left[ (2i-1) \ln u_i + (2(n-i)+1) \ln(1-u_i) \right].$$
 (7)

### III. DISTRIBUTION OF THE ANDERSON-DARLING STATISTIC

The asymptotic distribution of  $W_n^2$  was derived by Anderson and Darling [1952]. Lewis [1961] undertook the tabulation of  $F(z;n) = P(W_n^2 \le z)$  for n=1,2,...,8 and for incremental values of z over the interval [0.025, 8.000]. Lewis' table entries were computed using a Monte Carlo procedure to generate an empirical approximation  $F_m(z;n)$  to the distribution function F(z;n) based on m samples of size n. At that time, computational restrictions essentially limited the accuracy of the table entries to within 0.00326 of the true value.

Following an analogous procedure based on expression (7) and the observation that the  $U_1$  are distributed U[0,1] [Feller, 1966], the table appearing in Lewis' paper was recalculated using a Cray-2 supercomputer. Table 1 lists the reconstruction of Lewis' table, now accurate within 0.0005. Again, z ranges from 0.025 to 8.000 and for n = 1, 2, ..., 10. The column labeled " $\infty$ " contains the asymptotic values, rounded to four decimal places.

To obtain this increased accuracy, a Kolmogorov-type bound [Conover, 1980] was used to construct a 95% confidence band for the distribution function F(z;n). In general, the width of a  $(1-\alpha)100\%$  confidence band is equal to twice the value of the  $(1-\alpha)100\%$  quantile of the Kolmogorov

statistic  $K_m = \sup_{-\infty < x < \infty} \sqrt{m} |F_m(x) - F(x)|$ , where m is the number of Monte Carlo samples of size n used in the construction of  $F_m(x)$ . With n fixed, the 95% confidence band can be made arbitrarily small by a suitable choice for

TABLE 1. Quantiles of F(z;n) for n=1 thru 10 and  $\infty$ .

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0.00000	(0000) (0000) (0000) (0000) (0000) (0120) (0431)	0.0000 0.0000 0.0000 0.0000 0.00014 0.0065 0.0152 0.0152 0.0431	0.0000 0.0000 0.0000 0.0000 0.0005 0.0025 0.0025	0.0000 0.0000 0.0000 0.0000 0.0000 0.0029 0.0029	0.0000	O.CKNK;	CXXXX	(XXXX)	CKKKC
0.00000	.0000 (0000 (0000 (0000 (0000 (0120 (0120 (0431	0.0000 0.0000 0.0000 0.0000 0.00014 0.0065 0.0152 0.0152 0.0431	0.0000 0.0000 0.0000 0.0000 0.0005 0.0075 0.0076	0.0000 0.0000 0.0000 0.0000 0.0008 0.0008	0.0000				
0.0000	.0000 .0000 .0000 .0000 .0022 .0120 .0260	0.0000 0.0000 0.0000 0.0005 0.0005 0.0152 0.0274 0.0431	0.0000 0.0000 0.0003 0.0003 0.0075 0.0160 0.0183	0.0000 0.0000 0.0000 0.0002 0.00081 0.0166	0.000	0.0000	O.CKIKIO	0.0000	()()()()
0.0000	.0000 .0000 .0000 .0022 .0120 .0260	0.0000 0.0000 0.00014 0.0005 0.0152 0.0274 0.0431	0.0000 0.00003 0.00025 0.00075 0.01060 0.0160	0.0000 0.0000 0.0000 0.0029 0.0081		0.0000	()()()()()	0.0000	0.0000
0.0000	.0000 .0000 .0000 .0022 .0120 .0260	0.0000 0.0000 0.0014 0.0065 0.0152 0.0274 0.0431	0.0003 0.0003 0.0025 0.0075 0.0160 0.0283	0.0000 0.00029 0.0081 0.0166	0,000,0	0.0000	0.0000	0.0000	0.0000
0.0000	.0000 .0000 .0022 .0120 .0260	0.00014 0.0014 0.0065 0.0152 0.0274 0.0431	0.0003 0.0025 0.0075 0.0160 0.0283	0.0006 0.0029 0.0081 0.0166	0.0001	0.0001	0.0001	0.0001	0.0003
00000	.0000 .0022 .0120 .0260	0.0014 0.0065 0.0152 0.0274 0.0431	0.0025 0.0075 0.0160 0.0283	0.0029 0.0081 0.0166	0.0008	0.0000	0.0010	0.0010	0.0014
W.W.	.0022 .0120 .0260	0.0065 0.0152 0.0274 0.0431 0.0621	0.0075	0.0081	0.0033	0.0034	0.0036	0.0037	0.0042
0.0000	.0120 .0260 .0431	0.0152 0.0274 0.0431 0.0621	0.0160	0.0166	0.0085	0.0087	0.0089	0.0000	0.0096
0.0000	0260	0.0274 0.0431 0.0621	0.0283		0.0170	0.0173	0.0174	0.0176	0.0180
0.0008	(431	0.0431	0.0440	0.0288	0.0291	0.0294	0.0294	0.0296	0.0296
0.0209		0.0621		0.0444	0.0445	0.0448	0.0447	0.0449	0.0443
0.0586	.0625		0.0628	0.0629	0.0629	0.0631	0.0628	0.0629	0.0618
6980.0	.0840	0.0838	0.0842	0.0839	0.0838	0.0837	0.0833	0.0834	0.0817
0.1147	1.1071	0.1078	0.1073	0.1068	0.1065	0.1063	0.1058	0.1056	0.1036
0.1421	.1318	0.1330	0.1319	0.1311	0.1305	0.1303	0.1296	0.1295	0.1269
0.1691	.1579	0.1591	0.1574	0.1564	0.1555	0.1552	0.1547	0.1543	0.1513
0.1955	.1853	0.1856	0.1833	0.1821	0.1812	0.1808	0.1803	0.1797	0.1764
0.2217	.2136	0.2120	0.2096	0.2083	0.2072	0.2066	0.2061	0.2056	0.2019
0.2474	.2420	0.2384	0.2358	0.2344	0.2331	0.2325	0.2321	0.2315	0.2276
0.2727	.2693	0.2646	0.2618	0.2603	0.2589	0.2584	0.2579	0.2573	0.2532
0.2976	.2955	0.2902	0.2875	0.2859	0.2844	0.2839	0.2834	0.2828	0.2786
0.3221	.3209	0.3154	0.3128	0.3111	0.3095	0.3091	0.3085	0.3079	0.3036
0.3462	.3455	0.3398	0.3373	0.3356	0.3341	0.3337	0.3329	0.3325	0.3281
0.3700	.3693	0.3637	0.3612	0.3595	0.3580	0.3576	0.3569	0.3564	0.3520
0.3936	.3922	0.3870	0.3844	0.3827	0.3812	0.3809	0.3803	0.3798	0.3753
0.4166	.4143	0.4096	0.4069	0.4052	0.4038	0.4035	0.4029	0.4024	0.3980
0.4394	.4357	0.4314	0.4286	0.4270	0.4256	0.4254	0.4248	0.4244	0.4199
0.4618	.4563	0.4524	0.4495	0.4481	0.4467	0.4464	0.4459	0.4456	0.4412

TABLE 1. continued

						<b>E</b>					
Z	<b>-</b>	2	8	4	٧,	y	7	œ	6	10	8
0.7500	0.5522	0.5056	0.4953	0.4923	0.4803	0.4879	0.4867	0.4864	0.4859	0.4856	0.4815
0.8000	0.5821	0.5458	0.5317	0.5289	0.5261	0.5249	0.5238	0.5235	0.5231	0.5228	0.5190
0.8500	0.6091	0.5794	0.5652	0.5627	0.5601	0.5590	0.5581	0.5580	0.5574	0.5573	0.5537
0.9000	0.6337	0.6090	0.5962	0.5938	0.5916	0.5905	0.5897	0.5897	0.5891	0.5891	0.5858
0.9500	0.6563	0.6358	0.6247	0.6225	0.6206	0.6196	0.6190	0.6188	0.6184	0.6183	0.6154
1.0000	0.6771	0.6601	0.6511	0.6489	0.6473	0.64(A	0.6458	0.6457	0.6453	0.6454	0.6427
1.0500	0.6963	0.6823	0.6754	0.6732	0.6718	0.6712	90/90	0.6704	0.6701	0.6703	0.6680
1.1000	0.7141	0.7028	<i>116977</i>	0.6958	0.6944	0.6939	0.6934	0.6933	0.6931	0.6932	0.6912
1.1500	0.7306	0.7217	0.7182	0.7164	0.7153	0.7149	0.7145	0.7144	0.7142	0.7143	0.7127
1.2000	0.7459	0.7392	0.7367	0.7356	0.7345	0.7342	0.7339	0.7339	0.7337	0.7338	0.7324
1.2500	0.7603	0.7555	0.7538	0.7531	0.7522	0.7520	0.7518	0.7518	0.7517	0.7518	0.7508
1.3000	0.7737	0.7704	0.7696	0.7693	0.7686	0.7684	0.7683	0.7683	0.7684	0.7684	0.7677
1.3500	0.7862	0.7844	0.7842	0.7844	0.7838	0.7837	0.7836	0.7836	0.7837	0.7838	0.7833
1.4000	0.7980	0.7975	0.7977	0.7982	0.7978	0.7977	0.7978	0.7977	0.7979	0.7980	0.7978
1.4500	0.8090	0.8097	0.8103	0.8110	0.8108	0.8109	0.8109	0.8108	0.8110	0.8112	0.8111
1.5000	0.8194	0.8210	0.8221	0.8229	0.8228	0.8230	0.8230	0.8230	0.8232	0.8235	0.8235
1.5500	0.8291	0.8316	0.8330	0.8340	0.8340	0.8342	0.8343	0.8343	0.8345	0.8348	0.8350
1.6000	0.8382	0.8416	0.8431	0.8442	0.8444	0.8447	0.8448	0.8448	0.8450	0.8453	0.8457
1.6500	0.8469	0.8508	0.8526	0.8538	0.8541	0.8544	0.8546	0.8546	0.8549	0.8550	0.8556
1.7000	0.8550	0.8595	0.8615	0.8627	0.8631	0.8634	0.8636	0.8636	0.8640	0.8642	0.8648
1.7500	0.8626	0.8677	0.8698	0.8709	0.8715	0.8718	0.8720	0.8720	0.8725	0.8727	0.8734
1.8000	0.8698	0.8754	0.8775	0.8787	0.8793	0.8797	0.8798	0.8799	0.8803	0.8805	0.8814
1.8500	0.8767	0.8826	0.8847	0.8860	0.8866	0.8870	0.8872	0.8872	0.8877	0.8879	0.8888
1.9000	0.8830	0.8893	0.8915	0.8928	0.8934	0.8938	0.8940	0.8941	0.8945	0.8947	0.8957
1.9500	0.8891	0.8956	0.8979	0.8991	0.8997	0.9001	0.9004	0.9006	0.9008	0.9011	0.9021
2.0000	0.8948	0.9015	0.9038	0.9051	0.9058	0.9061	0.9064	0.9066	0.9068	0.9071	0.9082
2.0500	0.9002	0.9070	0.9093	0.9107	0.9113	0.9117	0.912C	0.9121	0.9124	0.9127	0.9138
2.1000	0.9053	0.9122	0.9146	0.9159	0.9165	0.9169	0.9172	0.9174	0.9177	0.9179	0.9190
2.1500	0.9102	0.9171	0.9195	0.9208	0.9214	0.9218	0.9221	0.9223	0.9226	0.9228	0.9239
2.2000	0.9148	0.9218	0.9241	0.9254	0.9259	0.9264	0.9266	0.9268	0.9272	0.9273	0.9285

TABLE 1. continued

						c					
2	1	2	3	4	5	9	7	œ	6	10	8
2.2500	0.9191	0.9262	0.9284	0.9297	0.9302	0.9307	0.9300	0.9311	0.9315	0.9316	0.9328
2.3000	0.9232	0.9303	0.9325	0.9337	0.9343	0.9347	0.9349	0.9351	0.9355	0.9356	0.9368
2.3500	0.9271	0.9342	0.9363	0.9375	0.9381	0.9385	0.9387	0.9389	0.9393	0.9394	0.9405
2.4000	0.9308	0.9378	0.9399	0.9411	0.9416	0.9420	0.9423	0.9425	0.9428	0.9429	0.9441
2.4500	0.9342	0.9412	0.9433	0.9444	0.9449	0.9454	0.9456	0.9458	0.9461	0.9463	0.9474
2.5000	0.9375	0.9445	0.9465	0.9476	0.9480	0.9485	0.9487	0.9489	0.9492	0.9494	0.9504
2.5500	0.9407	0.9475	0.9494	0.9505	0.9510	0.9514	0.9516	0.9519	0.9521	0.9523	0.9534
2.6000	0.9436	0.9504	0.9523	0.9532	0.9537	0.9542	0.9544	0.9546	0.9548	0.9550	0.9561
2.6500	0.9465	0.9531	0.9549	0.9558	0.9563	0.9567	0.9570	0.9572	0.9574	0.9576	0.9586
2.7000	0.9491	0.9556	0.9574	0.9583	0.9588	0.9592	0.9594	0.9596	0.9599	0.9600	0.9610
2.7500	0.9517	0.9580	0.9598	9096.0	0.9611	0.9614	0.9617	0.9619	0.9622	0.9623	0.9633
2.8000	0.9541	0.9602	0.9620	0.9628	0.9633	0.9636	0.9638	0.9640	0.9643	0.9644	0.9654
2.8500	0.9564	0.9624	0.9641	0.9648	0.9653	0.9657	0.9659	0.9661	0.9663	0.9665	0.9674
2.9000	0.9586	0.964	0.9661	0.9668	0.9672	0.9676	0.9678	0.9680	0.9682	0.9684	0.9692
2.9500	0.9607	0.9663	0.9679	0.9686	0.9690	0.9694	9696.0	0.9698	0.9700	0.9701	0.9710
3.0000	0.9626	0.9681	9696.0	0.9703	0.9708	0.9711	0.9713	0.9714	0.9717	0.9718	0.9726
3.0500	0.9645	0.9697	0.9713	0.9719	0.9724	0.9727	0.9729	0.9730	0.9733	0.9734	0.9742
3.1000	0.9663	0.9713	0.9728	0.9735	0.9739	0.9742	0.9744	0.9745	0.9748	0.9749	0.9756
3.1500	0.9679	0.9729	0.9743	0.9749	0.9753	0.9757	0.9758	0.9760	0.9762	0.9763	0.9770
3.2000	0.9695	0.9743	0.9757	0.9763	0.9767	0.9770	0.9771	0.9773	0.9775	0.9776	0.9783
3.2500	0.9711	0.9756	0.9770	9776.0	0.9780	0.9783	0.9784	0.9786	0.9787	0.9788	0.9795
3.3000	0.9725	0.9769	0.9782	0.9788	0.9792	0.9794	0.9796	0.9797	0.9799	0.9800	0.9807
3.3500	0.9738	0.9781	0.9794	0.9799	0.9803	0.9806	0.9807	0.9809	0.9810	0.9811	0.9818
3.4000	0.9751	0.9792	0.9805	0.9810	0.9814	0.9816	0.9817	0.9819	0.9821	0.9822	0.9828
3.4500	0.9764	0.9803	0.9815	0.9820	0.9824	0.9826	0.9827	0.9829	0.9830	0.9831	0.9837
3.5000	0.9775	0.9814	0.9825	0.9830	0.9833	0.9836	0.9837	0.9839	0.9840	0.9840	0.9846
3.5500	0.9786	0.9823	0.9834	0.9839	0.9842	0.9844	0.9846	0.9847	0.9849	0.9849	0.9855
3.6000	0.9797	0.9833	0.9843	0.9848	0.9851	0.9853	0.9854	0.9856	0.9857	0.9857	0.9863
3.6500	0.9807	0.9841	0.9851	0.9856	0.9859	0.9861	0.9862	0.9864	0.9865	0.9865	0.9870
3.7000	0.9816	0.9849	0.9859	0.9863	0.9867	0.9868	0.9869	0.9871	0.9872	0.9872	0.9878

TABLE 1. continued

						c					
2		2	3	4	s	9	7	œ	6	10	8
3.7500	0.9825	0.9857	1986.0	0.9871	0.9874	0.9875	0.9877	0.9878	0.0879	0.9879	0.9884
3.8000	0.9834	0.9865	0.9874	0.9878	0.9881	0.9882	0.9883	0.9885	0.9886	0.9886	0.9891
3.8500	0.9842	0.9872	0.9880	0.9884	0.9887	0.9888	0.9880	0.9891	0.9892	0.9892	0.9897
3.9000	0.9850	0.9878	0.9887	0.9890	0.9893	0.9894	0.9895	0.9847	0.9898	0.9898	0.9902
3.9500	0.9858	0.9885	0.9893	9686.0	0.9899	0.9900	0.9901	0.9902	0.9903	0.9903	0.9908
4.0000	0.9864	0.9890	0.9898	0.9902	0.9904	0.9905	0.9906	0.9908	0.9909	0.9908	0.9913
4.0500	0.9871	0.9896	0.9904	0.9907	0.9909	0.9910	0.9911	0.9913	0.9913	0.9913	0.9917
4.1000	0.9877	0.9901	6066.0	0.9912	0.9914	0.9915	0.9916	0.9917	0.9918	0.9918	0.9922
4.1500	0.9883	9066.0	0.9914	0.9917	0.9919	0.9919	0.9920	0.9922	0.9922	0.9922	0.9926
4.2000	0.9889	0.9911	0.9918	0.9921	0.9923	0.9924	0.9925	0.9926	0.9927	0.9927	0.9930
4.2500	0.9895	0.9916	0.9922	0.9925	0.9927	0.9928	0.9929	0.9930	0.9931	0.9931	0.9934
4.3000	0.9900	0.9920	0.9926	0.9929	0.9931	0.9931	0.9932	0.9934	0.9934	0.9934	0.9938
4.3500	0.9905	0.9924	0.9930	0.9933	0.9935	0.9935	0.9936	0.9937	0.9938	0.9938	0.9941
4.4000	0.9910	0.9928	0.9934	0.9936	0.9938	0.9938	0.9939	0.9941	0.9941	0.9941	0.9944
4.5000	0.9918	0.9935	0.9941	0.9943	0.9945	0.9945	0.9946	0.9947	0.9948	0.9947	0.9950
4.6000	0.9926	0.9941	0.9946	0.9949	0.9950	0.9951	0.9951	0.9952	0.9953	0.9953	0.9955
4.7000	0.9933	0.9947	0.9952	0.9954	0.9956	0.9956	0.9956	0.9957	0.9958	0.9958	0.9960
4.8000	0.9939	0.9952	0.9957	0.9959	0.9960	0.9960	0.9961	0.9962	0.9962	0.9962	0.9964
4.9000	0.9945	0.9957	0.9961	0.9963	0.9964	0.9964	0.9965	99660	0.9966	9966	0.9968
5.0000	0.9950	0.9961	0.9965	0.9967	0.9968	0.9968	0.9968	0.9969	0.9969	0.9969	0.9971
2.5000	0.9970	0.9977	0.9979	0.9980	0.9981	0.9981	0.9982	0.9982	0.9982	0.9982	0.9983
0000.9	0.9982	0.9986	0.9988	0.9988	0.9989	0.9989	0.9989	0.9989	0.666.0	0.9990	0.9990
7.0000	0.9993	0.9995	9666:0	9666.0	9666.0	96660	9666'0	9666.0	0.9996	9666.0	0.9997
8.0000	0.9997	0.9998	0.9998	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999

m. The commonly tabled [Miller, 1956] asymptotic approximation for the 95<sup>th</sup> quantile is 1.358/Vm. However, Harter [1980] suggests using

$$\frac{1.358}{\left[m + \frac{r}{3.5}\right]^{\frac{1}{2}}}, \quad \text{where } r = (m+4)^{\frac{1}{2}}$$
 (8)

for an improved approximation.

Using approximation (8) to construct a 95% confidence band with the width not exceeding 0.001, the value for m must be at least 7,375,881. In this simulation, m was chosen to be 7.4 million<sup>1</sup>.

### IV. POWER STUDY

The power of the Anderson-Darling test was compared with two other goodness-of-fit procedures based on the empirical distribution function: the Kolmogorov and the Cramér-von Mises tests. The Kolmogorov statistic introduced in Section 1 as metric (2) with weighting function  $\psi$   $[F(x)] \equiv 1$  becomes

$$K_{n} = \sup_{-\infty < x < \infty} \sqrt{n} |F_{n}(x) - F(x)|.$$
 (9)

For an ordered sample  $x_1 \le x_2 \le \cdots \le x_n$  and  $F(x_i) = u_i$ ,  $K_n$  may be evaluated as  $\sqrt{n}D$  where  $D = max(D^+, D^-)$  and

$$D^{+} = \max_{i} \left( \frac{i}{n} - u_{i} \right),$$

$$D^{-} = \max_{i} \left\{ u_{i} - \frac{i-1}{n} \right\}.$$

The sample values were obtained via a linear congruential uniform random number generator of the form  $X_{k+1} = (aX_k + c)_{mod q}$ , where  $a = 273673163155_8$ ,  $c = 13_8$ , and  $q = 2^{48}$ . It has the properties of a 'good' random number generator as suggested by Rubinstein [1981]. In addition, several subsets of the random numbers generated were tested for autocorrelations with lag up to 36. For each subset and each lag, the autocorrelation did not exceed 0.06 in absolute value.

The Cramér-von Mises statistic, defined as

$$W_n^2 = n \int_{-\infty}^{\infty} [F_n(x) - F(x)]^2 dF(x)$$

can be reduced to (10) for ease of computation (Appendix C);

$$W_n^2 = \sum_{i=1}^n \left[ u_i - \frac{2i-1}{2n} \right]^2 + \frac{1}{12n} . \tag{10}$$

In the power study, two cases were considered. Case 1 corresponds to the situation in which the parameters of the hypothesized distribution are completely specified. Case 2 corresponds to the situation in which the parameters are not specified and must be estimated from the sample data.

For both case 1 and 2, the null hypothesis is

 $H_c$ : A random sample  $X_1, X_2, ..., X_n$  comes from a normal population or  $H_c$ :  $F(x) = F_c(x)$ , where  $F_c(x) \sim N(\mu, \sigma^2)$ .

As alternative hypotheses, the Cauchy, double exponential, and extreme value distributions were chosen, each with location parameter the same as the null hypothesis. This provided a heavy-tailed, light-tailed, and skewed distribution, respectively, against which the power of the three goodness-of-fit tests are compared.

The power functions do not exist in closed form; they are approximated empirically via a Monte Carlo simulation. To determine a point on the power curve, a large number of samples of size n was generated from a specific distribution serving as the alternative hypothesis. The number of times that the null hypothesis was rejected at a specific level of significance was recorded. The ratio of the number of rejections, Y, to the total number of samples generated. N, provides an estimate,  $\hat{p} = Y/N$ , of the probability of rejecting the null hypothesis when it should be rejected (power). The value  $\hat{p}$  determines a point on the power curve corresponding to a specific sample size n, a specific significance level  $\alpha$ , and a specific alternative hypothesis.

To determine the number of samples of size n required for a sufficiently accurate estimate of  $\hat{p}$ , a nonparametric technique was employed. Since the counter Y is distributed binomial(p, N) where the parameter p is the true but unknown power, and since an approximate confidence interval for p can be constructed [Conover, 1980] using

$$1 - \alpha \simeq P\left(\frac{Y}{N} - z_{(1 - \alpha/2)} \left\{ \frac{Y}{N} (1 - \frac{Y}{N})/N \right\}^{\frac{1}{2}}$$

samples of size n continued to be generated from the alternative distribution until the confidence interval for p given in (11) was sufficiently small.

The confidence interval coefficient  $1 - \alpha$  was chosen to be 0.975 and the confidence interval width not to exceed 0.025. From (11), this will occur when

the inequality 
$$2 \cdot z_{0.9875} \left\{ \frac{Y}{N} (1 - \frac{Y}{N}) / N \right\}^{\frac{1}{2}} \le 0.025$$
 is satisfied. For any value of

N. the left-hand side of the inequality is maximum when Y/N is 1/2. Substituting Y/N = 1/2 into the inequality yields N  $\leq 8037$  — the largest possible number of samples required. In practice, the value for N will usually be much smaller and will change for each estimate  $\hat{p}$ . This dynamic scheme was chosen rather than fixing N = 8037 for the entire simulation. A minimum value for N of 100 was imposed to prevent premature termination of the procedure.

### 1. Case 1: Distribution Parameters Specified.

The power study for case 1 specified the parameters of the hypothesized distribution as N(0.1). The results of the study are summarized in Figures 1-12. For each of the three distributions serving as an alternative hypothesis, samples of size n=5, 10, 15, 20 were chosen for study and, as previously mentioned, the location parameters of both the null and alternative hypotheses coincided. The scale parameter for the alternative hypothesis ranged from 0.025 to 3.000 in increments of 0.025.

The level of significance for the study was 0.05. The critical value for each test was determined from tables in Conover [1980] for the Kolmogorov test. Stephens and Maag [1968] for the Cramér-von Mises test, and Table 1 in Section III of this paper for the Anderson-Darling test.

The Anderson-Darling test demonstrated overall superiority for the sample sizes and hypotheses chosen for this study. This is perhaps to be anticipated in view of the emphasis on agreement in the tails by the Anderson-Darling procedure, but the magnitude of difference over the Kolmogorov and Cramérvon Mises tests is impressive.

The power curves corresponding to n = 10, 15, 20 are distinguished by their characteristic of decreasing to a global minimum before becoming

monotonically increasing. An explanation of this feature is suggested by consideration of Figures 13 – 15 in which the distribution functions of the N(0,1) and Cauchy  $(0,\varsigma)$  are compared. There it is seen (Figure 14) that corresponding to  $\varsigma=0.50$  the two distribution functions are similar; an increase (decrease) in the scale parameter  $\varsigma$  causes the tails of the distributions to become more distinct. Values in a neighborhood of  $\varsigma\simeq0.50$  marked the global minimum throughout the study.

### 2. Case 2: Distribution Parameters Estimated.

The Anderson-Darling, Kolmogorov, and Cramér-von Mises goodness-of-fit tests were developed for use in case 1 where distribution parameters are specified, and so precludes their use in the more likely situation where parameters must be estimated. In practice, these procedures are sometimes used anyway with the caveat that the tests are likely to be conservative. Stephens [1974] provides adjustments to the test statistics that enables the tests to be used to test the assumption  $H_o$ :  $F(x) = F_o(x)$ , where  $F_o(x) \sim N(\mu, \sigma^2)$  and the population parameters are estimated from the data.

The results of the power study for case 2, are summarized in Figures 16-27. As in case 1, the sample sizes are n=5, 10, 15, and 20, and the level of significance is 0.05. Both location and scale parameters coincide: the scale parameter are values from 0.025 to 3.000 in increments of 0.025.

The power plots are horizontal, demonstrating that power does not change with scale parameter and provides empirical support for Stephens' transformations. Power increases with increasing sample size, as would be expected. When n = 5, the Anderson-Darling test was the least powerful of all three distributions examined. However, none of the distributions had power above 0.30 for this sample size. At the larger sample sizes, the Anderson-Darling test was only slightly better than the Kolmogorov and the Cramér-von Mises tests. In general, when both location and scale parameters agree, all three tests are competitive for the sample sizes and alternative distributions chosen for this study.

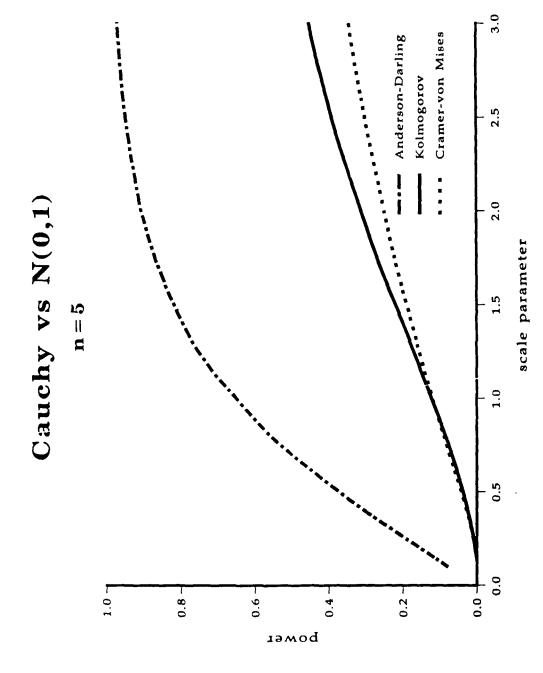
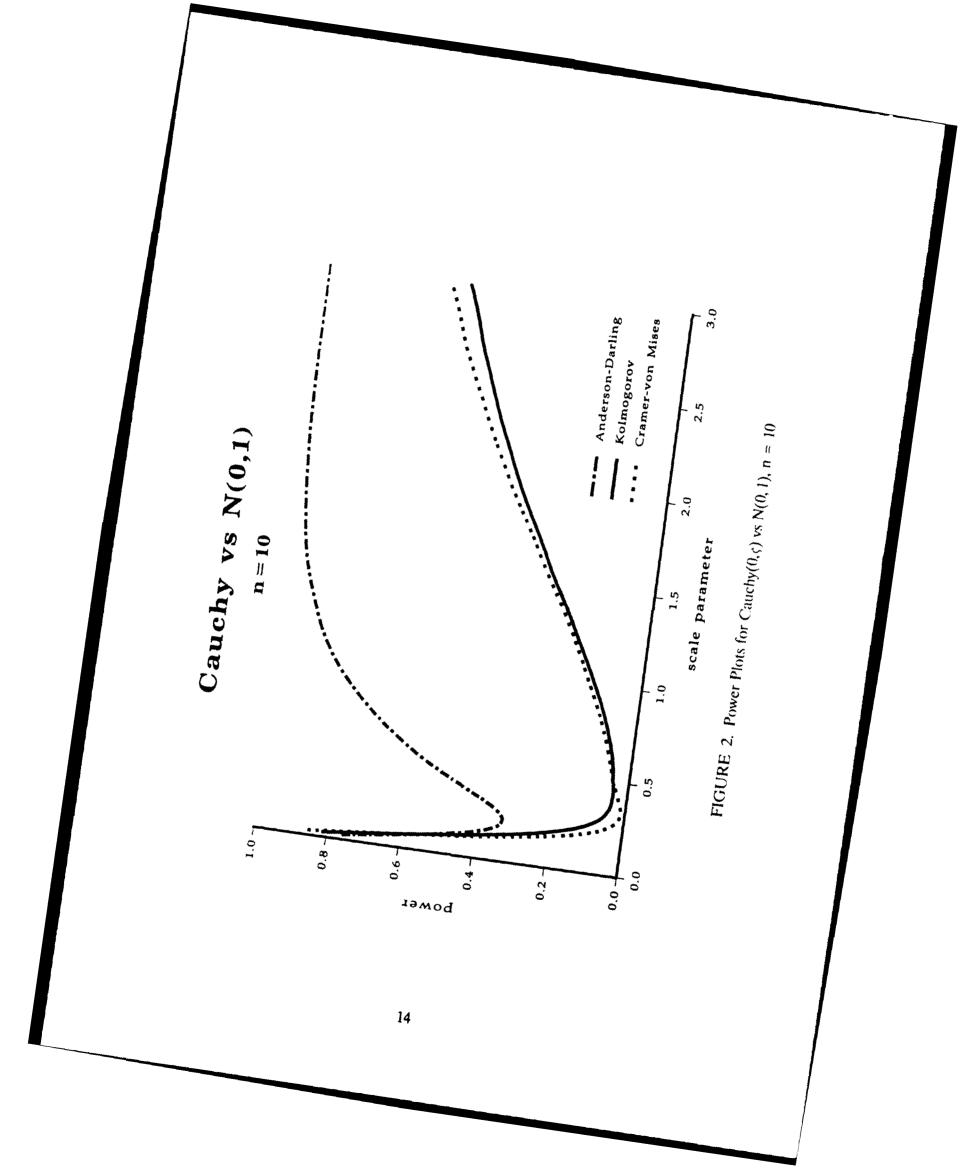
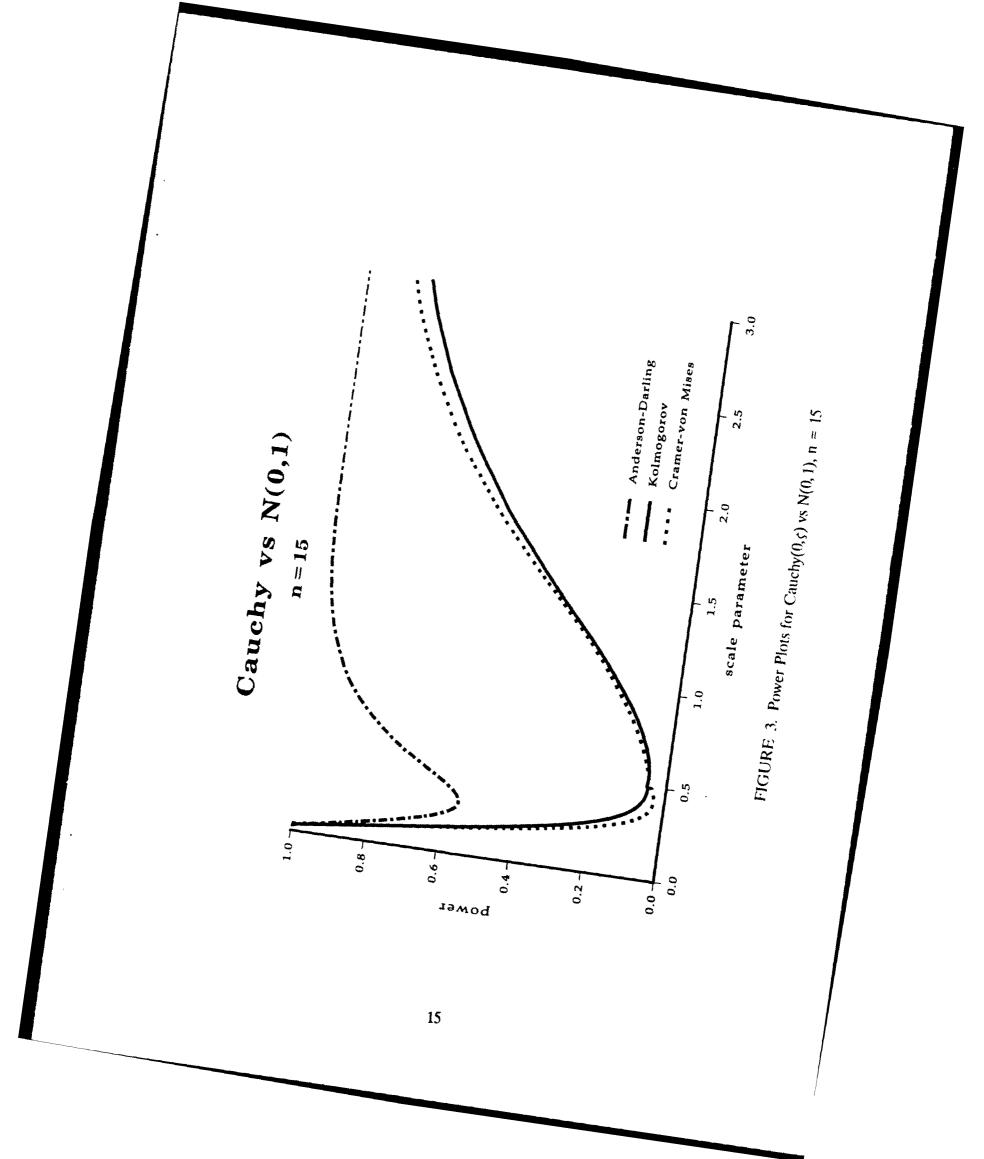
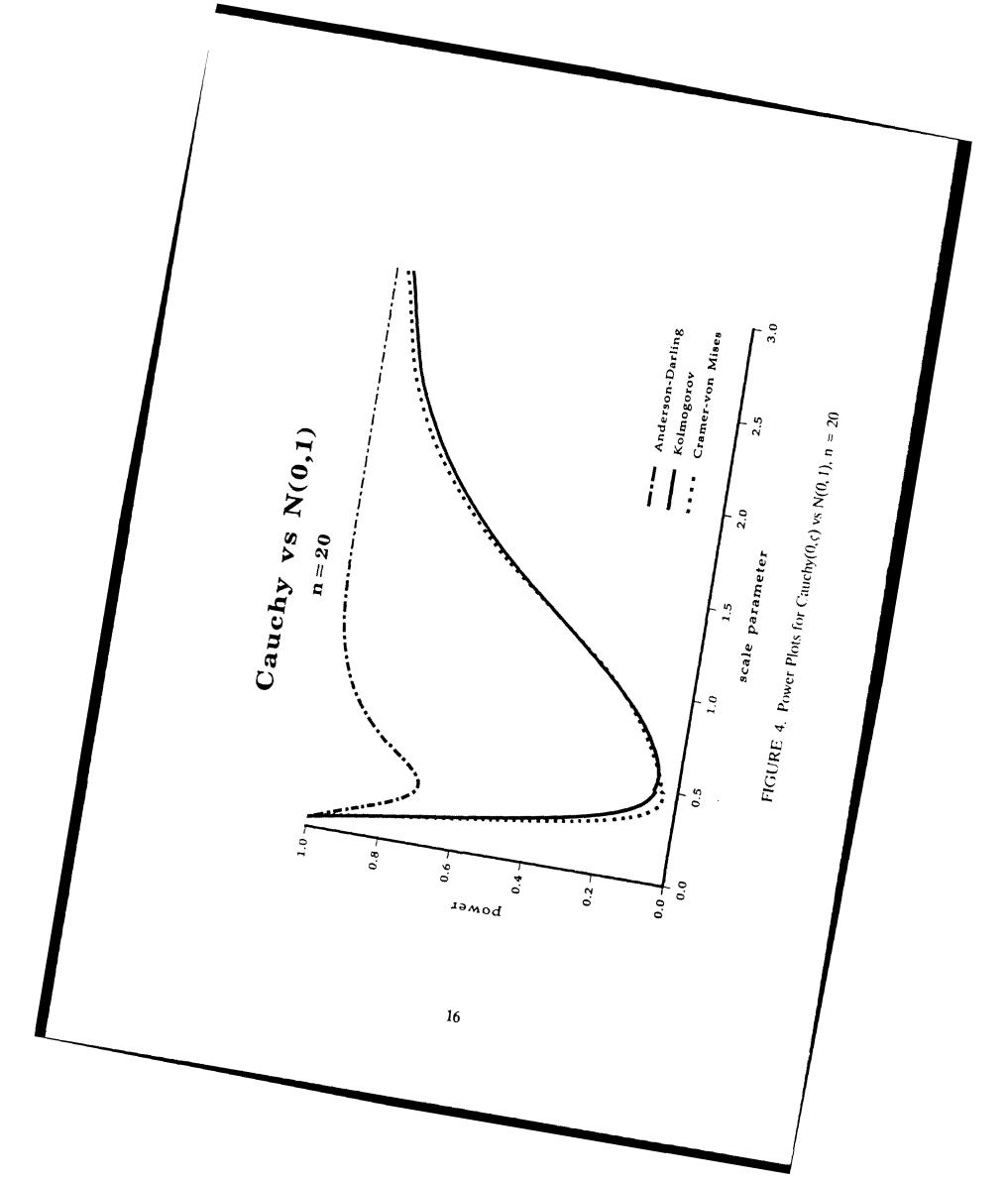


FIGURE 1. Power Plots for Cauchy(0, $\zeta$ ) vs N(0,1), n = 5







# Double Exponential vs N(0,1)

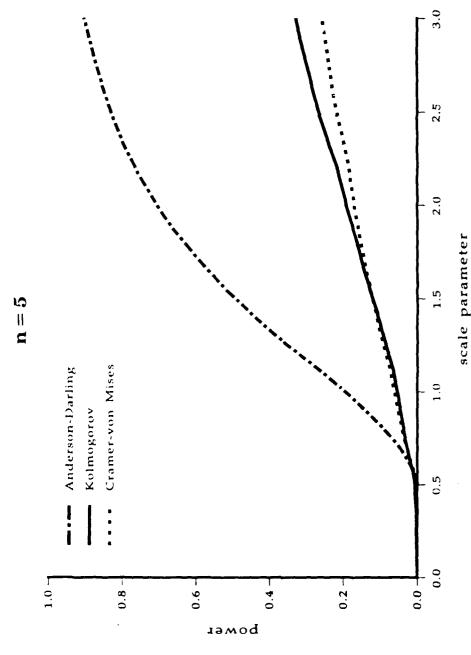
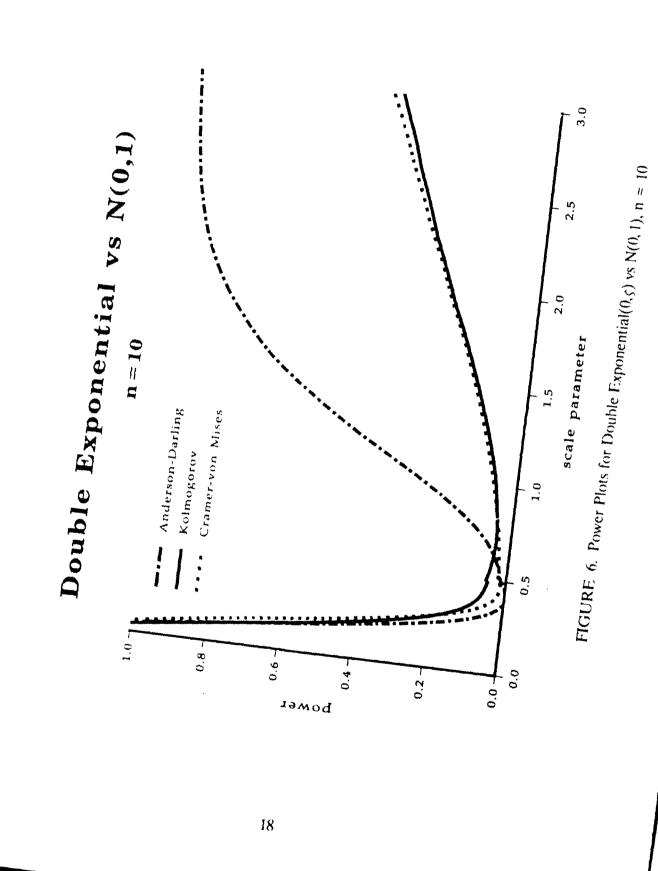


FIGURE 5. Power Plots for Double Exponential(0, $\xi$ ) vs N(0,1), n = 5



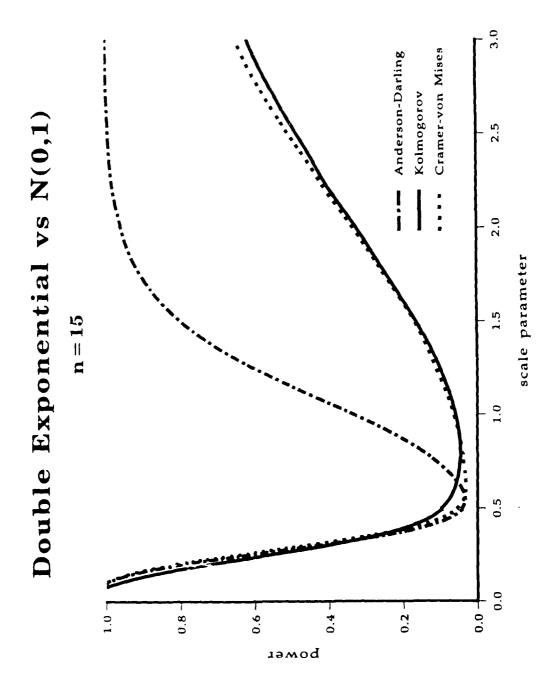


FIGURE 7. Power Plots for Double Exponential(0, $\varsigma$ ) vs N(0,1), n = 15

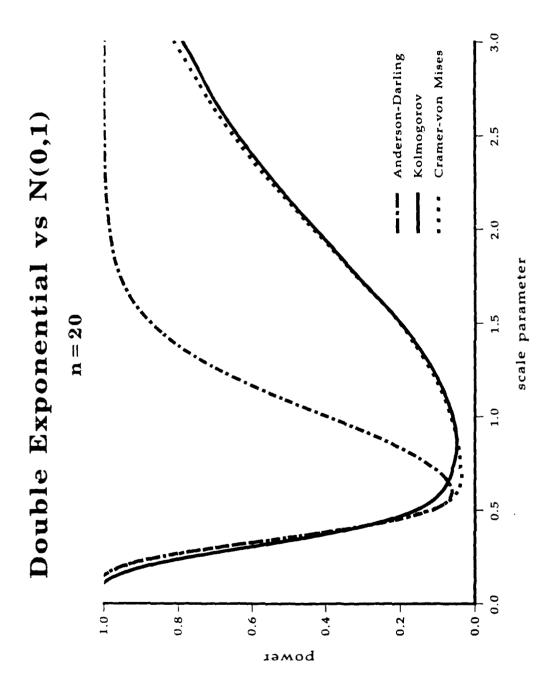


FIGURE 8. Power Plots for Double Exponential(0, $\xi$ ) vs N(0,1), n = 20

## Extreme Value vs N(0,1)

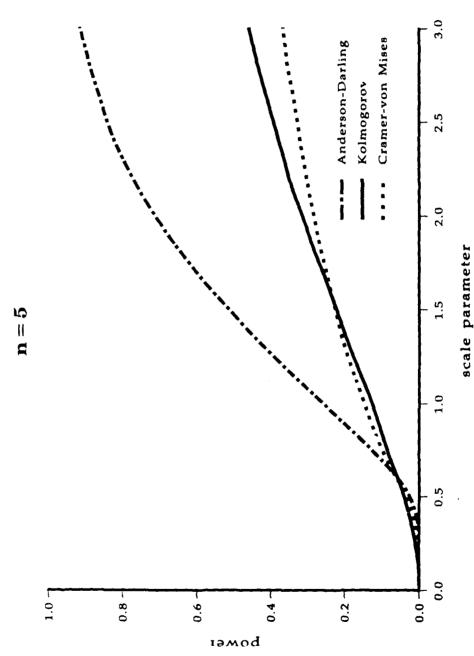


FIGURE 9. Power Plots for Extreme Value(0, $\varsigma$ ) vs N(0,1), n = 5

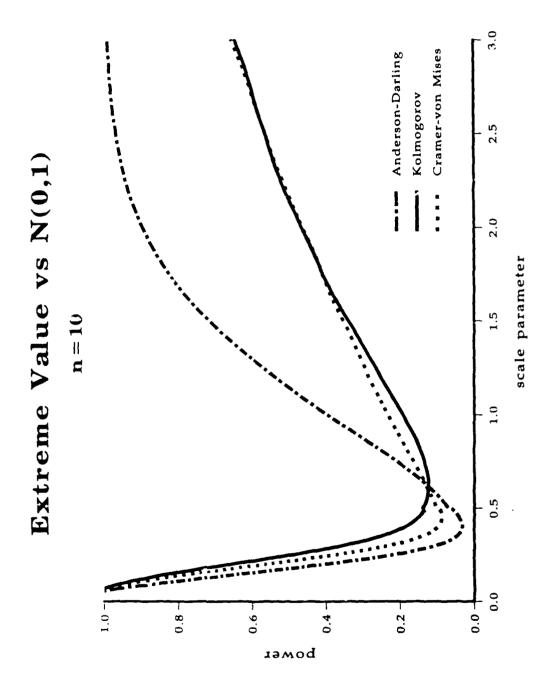


FIGURE 10. Power Plots for Extreme Value(0, $\varsigma$ ) vs N(0,1), n = 10

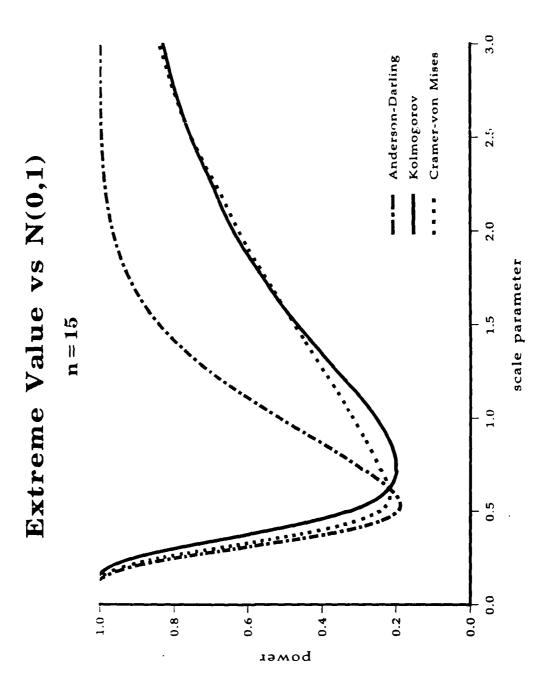


FIGURE 11. Power Plots for Extreme Value(0, $\zeta$ ) vs N(0,1), n = 15

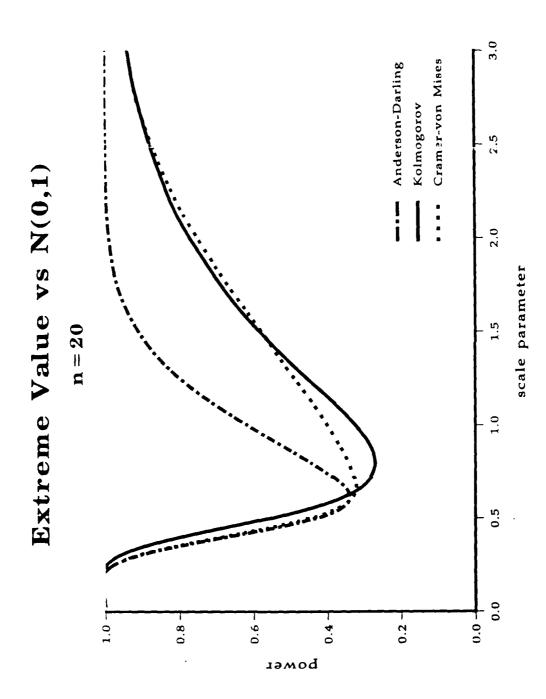


FIGURE 12. Power Plots for Extreme Value(0, $\zeta$ ) vs N(0,1), n = 20

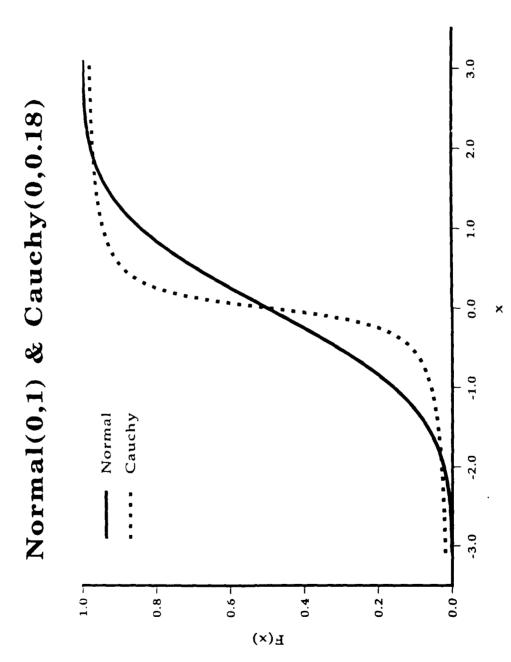


FIGURE 13. Cumulative Distribution Curves, Normal(0,1) & Cauchy(0,0.18)



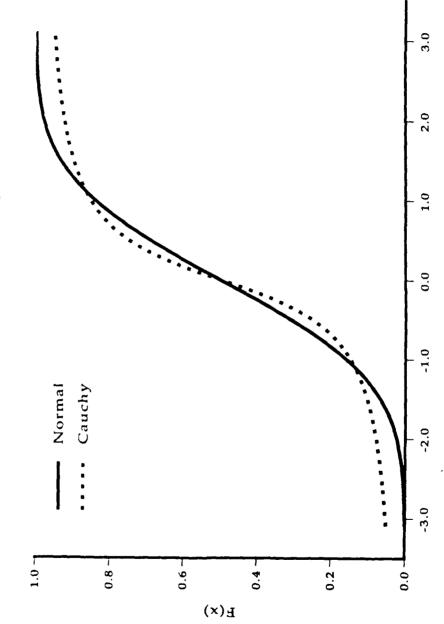


FIGURE 14. Cumulative Distribution Curves, Normal(0,1) & Cauchy(0,0.50)

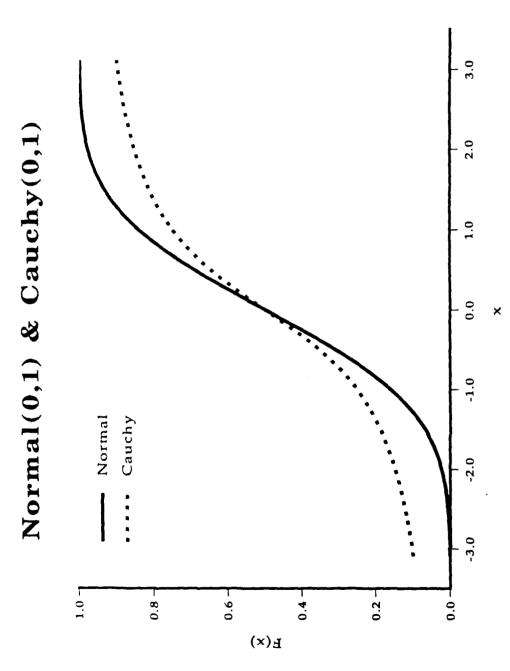


FIGURE 15. Cumulative Distribution Curves, Normal(0,1) & Cauchy(0,1)

### Cauchy vs N(x, s2)

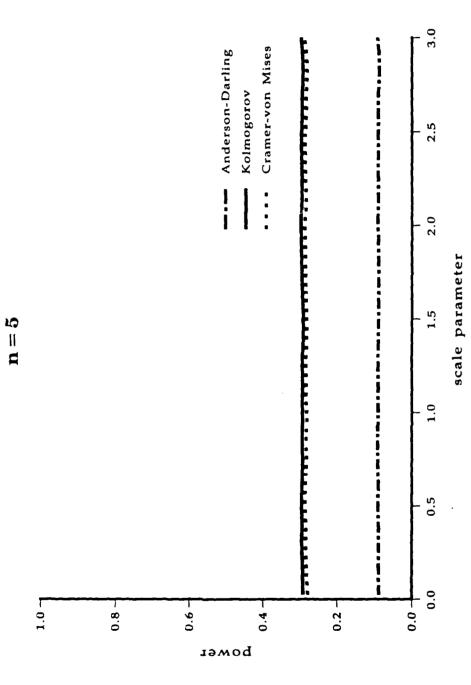


FIGURE 16. Power Plots for Cauchy(0, $\varsigma$ ) vs N( $\bar{x}$ , $s^2$ ), n = 5

### Cauchy vs N(x, s2)



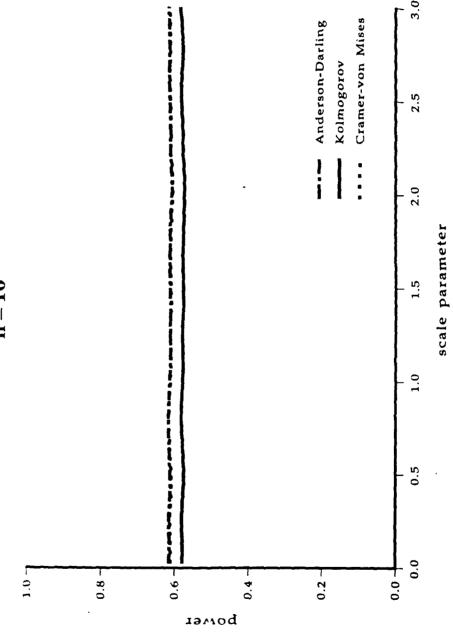


FIGURE 17. Power Plots for Cauchy(0, $\zeta$ ) vs N( $\bar{x}$ ,  $s^2$ ),  $n \approx 10$ 

### Cauchy vs N(x, s2)



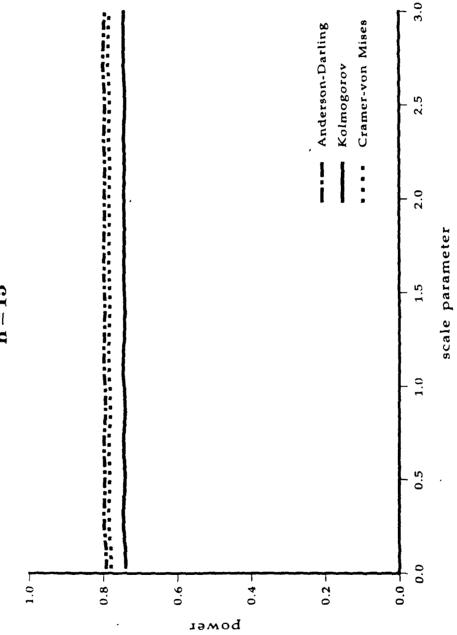


FIGURE 18. Power Plots for Cauchy(0, $\varsigma$ ) vs N( $\bar{x}$ , $s^2$ ), n = 15

### Cauchy vs N(x, s2)

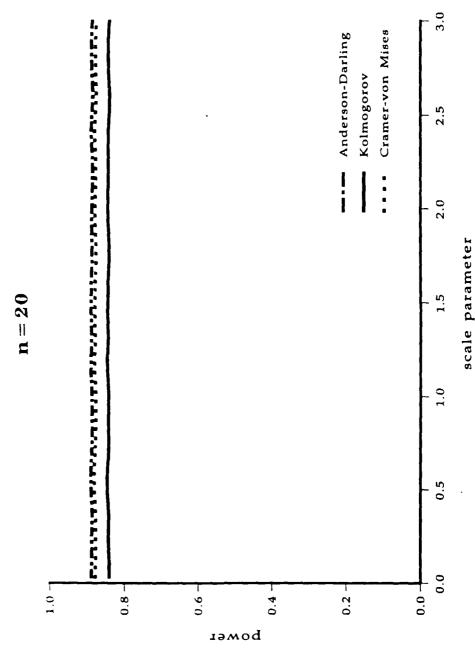


FIGURE 19. Power Plots for Cauchy(0, $\varsigma$ ) vs N( $\bar{x}$ ,  $s^2$ ), n = 20

# Double Exponential vs N(x, s2)

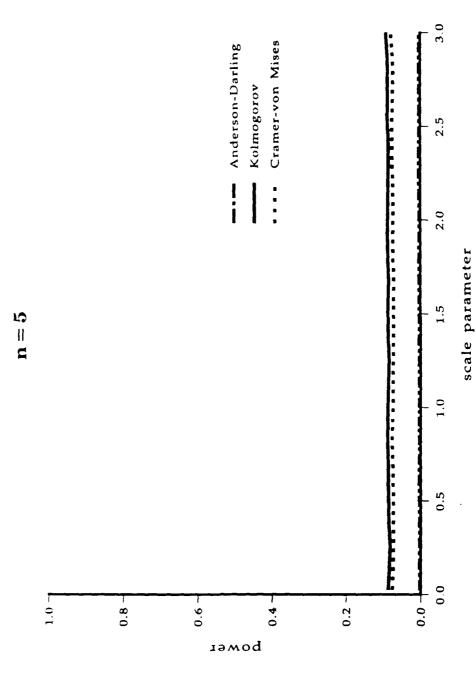


FIGURE 20. Power Plots for Double Exponential(0,  $\varsigma$ ) vs N( $\bar{x}$ ,  $s^2$ ), n = 5

# Double Exponential vs N(x, s2)

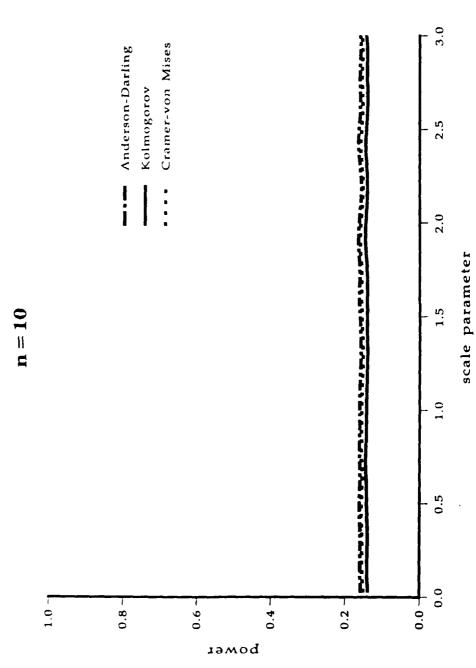


FIGURE 21. Power Plots for Double Exponential(0, $\zeta$ ) vs N( $\bar{x}$ , s<sup>2</sup>), n = 10

# Double Exponential vs N(x̄, s²)

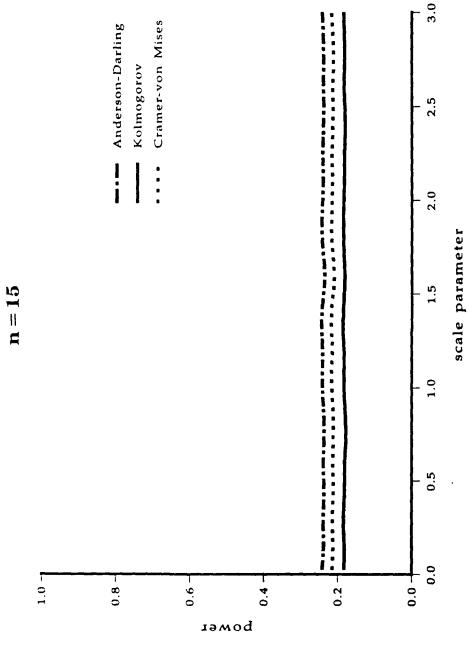


FIGURE 22. Power Plots for Double Exponential(0, $\xi$ ) vs N( $\bar{x}$ , s<sup>2</sup>), n = 15

# Double Exponential vs N(x, s2)

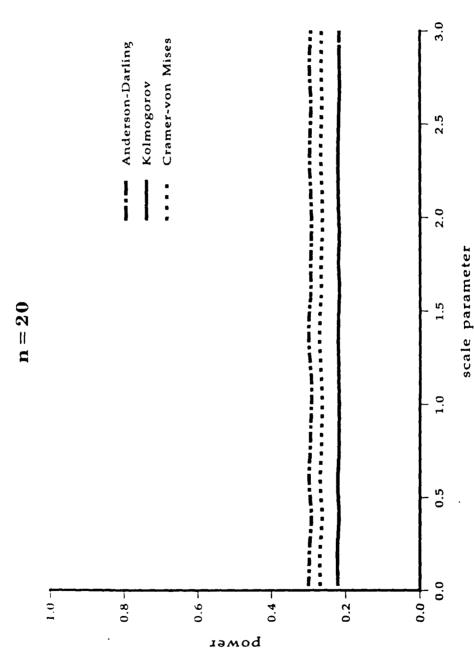


FIGURE 23. Power Plots for Double Exponential(0, $\zeta$ ) vs N( $\bar{x}$ , s<sup>2</sup>), n = 20

## Extreme Value vs N(x, s2)

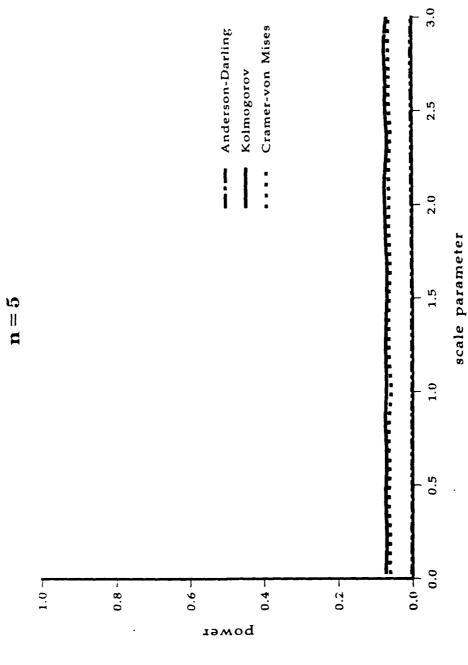


FIGURE 24. Power Plots for Extreme Value(0, $\xi$ ) vs N( $\bar{x}$ , $s^2$ ), n = 5

## Extreme Value vs N(x, s²)

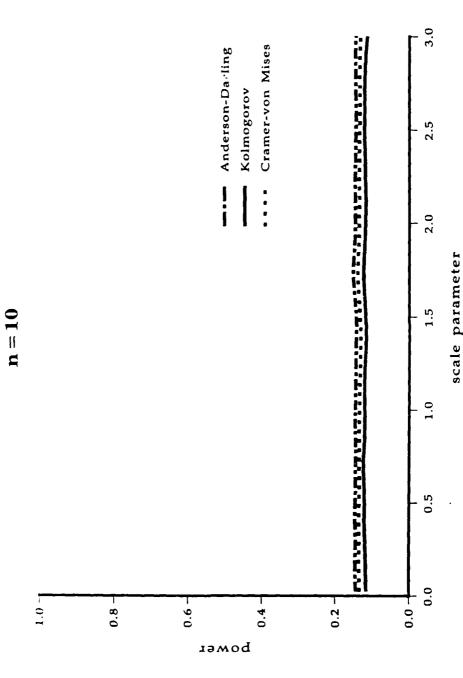


FIGURE 25. Power Plots for Extreme Value(0, $\xi$ ) vs N( $\bar{x}$ , s<sup>2</sup>), n = 10

## Extreme Value vs N(x, s2)

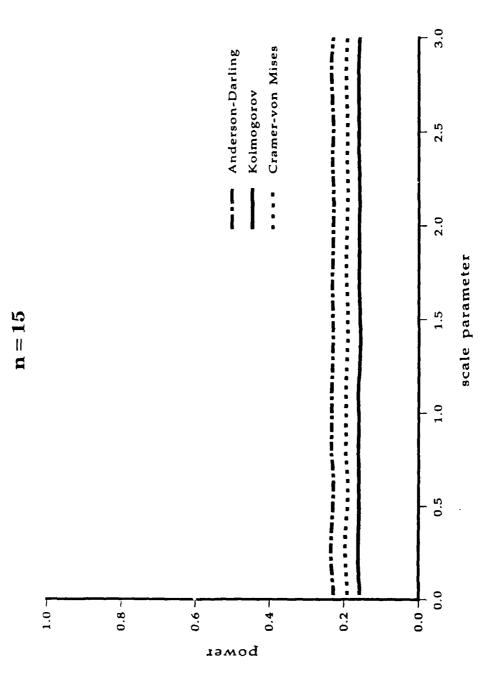
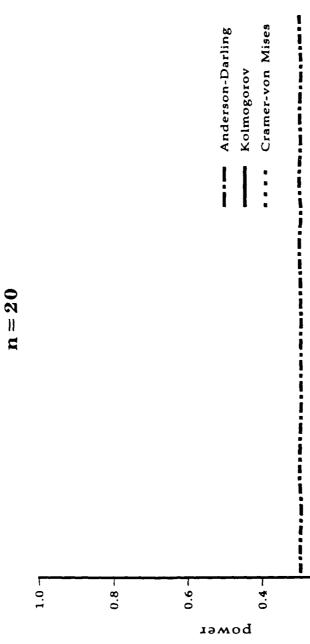


FIGURE 26. Power Plots for Extreme Value(0, $\varsigma$ ) vs N( $\bar{x}$ , $s^2$ ), n = 15

## Extreme Value vs N(x, s2)



0.2

FIGURE 27. Power Plots for Extreme Value(0, $\zeta$ ) vs N( $\bar{x}$ , s<sup>2</sup>), n = 20

scale parameter

2.5

2.0

0.5

0.0

0.0

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APPENDIX A: EXPECTATION OF SQUARED DISCREPANCY BETWEEN AN EMPIRICAL DISTRIBUTION FUNCTION AND A SPECIFIED DISTRIBUTION FUNCTION

### APPENDIX A: EXPECTATION OF SQUARED DISCREPANCY BETWEEN AN EMPIRICAL DISTRIBUTION FUNCTION AND A SPECIFIED DISTRIBUTION FUNCTION

$$\begin{split} & E\left[F_{n}(x) - F(x)\right]^{2} = E\left[F_{n}(x) - H(x) + H(x) - F(x)\right]^{2} \\ & = E\left[\left\{F_{n}(x) - H(x)\right\} - \left\{F(x) - H(x)\right\}\right]^{2} \\ & = E\left[\left\{F_{n}(x) - H(x)\right\}^{2} - 2\left\{F_{n}(x) - H(x)\right\}\left\{F(x) - H(x)\right\} + \left\{F(x) - H(x)\right\}^{2} \right] \\ & = E\left[F_{n}(x) - H(x)\right]^{2} - 2\left\{F(x) - H(x)\right\} E\left[F_{n}(x) - H(x)\right] \\ & + \left\{F(x) - H(x)\right\}^{2} \\ & = E\left[F_{n}(x) - H(x)\right]^{2} + \left\{F(x) - H(x)\right\}^{2} \\ & = E\left[F_{n}^{2}(x) - 2F_{n}(x)H(x) + H^{2}(x)\right] + \left\{F(x) - H(x)\right\}^{2} \\ & = E\left[F_{n}^{2}(x)\right] - 2H(x) E\left[F_{n}(x)\right] + H^{2}(x) + \left\{F(x) - H(x)\right\}^{2} \end{split}$$

$$= \left[\frac{1}{n} H(x)\{1 - H(x)\} + H^{2}(x) - 2 H^{2}(x) + H^{2}(x) + \left\{F(x) - H(x)\right\}^{2}\right]$$

$$+ \left\{F(x) - H(x)\right\}^{2}$$

$$+ \left\{F(x) - H(x)\right\}^{2} - \left[\frac{1}{n} \left\{H(x)\{1 - H(x)\}\right\} + \left\{F(x) - H(x)\right\}^{2}\right].$$

APPENDIX B: EXPANSION AND INTEGRATION OF THE ANDERSON-DARLING STATISTIC

### APPENDIX B: EXPANSION AND INTEGRATION OF THE ANDERSON-DARLING STATISTIC

$$W_n^2 = n \int_{-\infty}^{\infty} \frac{[F_n(x) - F(x)]^2}{F(x)[1 - F(x)]} dF(x)$$

$$= n \left\{ \int_{-\infty}^{x_1} \frac{[F_n(x) - F(x)]^2}{F(x)[1 - F(x)]} dF(x) + \sum_{k=1}^{n-1} \int_{x_k}^{x_{k+1}} \frac{[F_n(x) - F(x)]^2}{F(x)[1 - F(x)]} dF(x) + \int_{x_n}^{\infty} \frac{[F_n(x) - F(x)]^2}{F(x)[1 - F(x)]} dF(x) \right\}$$

Let 
$$F(x) = u$$
  
 $dF(x) = du$   
 $F(x_i) = u_i$ 

$$W_{n}^{2} = n \begin{cases} u_{1} & \frac{1}{\left[0 - u\right]^{2}} du + \sum_{k=1}^{n-1} \int_{u_{k}}^{u_{k+1}} \frac{\left(\frac{k}{n} - u\right)^{2}}{u(1 - u)} du + \int_{u_{n}}^{1} \frac{(1 - u)^{2}}{u(1 - u)} du \end{cases}$$

$$= n \begin{cases} -u_{1} - \ln(1 - u_{1}) + \sum_{k=1}^{n-1} \int_{u_{k}}^{u_{k+1}} \frac{\left(\frac{k^{2}}{n^{2}} - \frac{2k}{n} u + u^{2}\right)}{u(1 - u)} du \\ + \left[-1 + u_{n} - \ln u_{n}\right] \end{cases}$$

$$= n \left\{ -u_1 - \ln (1 - u_1) + \sum_{k=1}^{n-1} \left[ \frac{k^2}{n^2} \int_{u_k}^{u_{k+1}} \frac{du}{u (1 - u)} \right] \right\}$$

$$+ \sum_{k=1}^{n-1} \left[ -\frac{2k}{n} \int_{u_k}^{u_{k+1}} \frac{u}{u(1-u)} du \right]$$

$$+ \sum_{k=1}^{n-1} \left[ \int_{u_k}^{u_{k+1}} \frac{u^2}{u(1-u)} du \right]$$

$$+ \left[ -1 + u_n - \ln u_n \right]$$

$$= n \left\{ -u_1 - \ln (1 - u_1) + \sum_{k=1}^{n-1} \frac{k^2}{n^2} \left[ \ln (u_{k+1}) - \ln (1 - u_{k+1}) - \ln u_k + \ln (1 - u_k) \right] \right.$$

$$+ \sum_{k=1}^{n-1} \frac{2k}{n} \left[ \ln (1 - u_{k+1}) - \ln (1 - u_k) \right]$$

$$+ \sum_{k=1}^{n-1} \left[ -u_{k+1} - \ln (1 - u_{k+1}) + u_k + \ln (1 - u_k) \right]$$

$$+ \left[ -1 + u_n - \ln u_n \right] \right\}$$

$$= n \left\{ -u_1 - \ln (1 - u_1) + \sum_{k=1}^{n-1} - \left( \frac{2k - 1}{n^2} \right) \left[ \ln u_k - \ln (1 - u_k) \right] \right.$$

$$+ \left. \left( \frac{n - 1}{n} \right)^2 \left[ \ln u_n - \ln (1 - u_n) \right] \right.$$

$$+ \left. \sum_{k=1}^{n-1} \left[ -\frac{2}{n} \ln (1 - u_k) \right] \right.$$

$$+ \left. \left( \frac{2n - 2}{n} \right) \ln (1 - u_n) \right.$$

$$+ \left. u_1 + \ln (1 - u_1) - u_n - \ln (1 - u_n) \right.$$

$$+ \left. \left[ -1 + u_n - \ln u_n \right] \right\}$$

$$= n \left\{ -1 - \sum_{k=1}^{n} \left[ \left( \frac{2k - 1}{n^2} \right) \ln u_k + \left( \frac{2(n - k) + 1}{n^2} \right) \ln (1 - u_k) \right] \right.$$

$$= -n - \frac{1}{n} \sum_{k=1}^{n} \left[ (2k - 1) \ln u_k + (2(n - k) + 1) \ln (1 - u_k) \right] .$$

APPENDIX C: DERIVATION OF THE CRAMÉR-VON MISES STATISTIC

### APPENDIX C: DERIVATION OF THE CRAMÉR-VON MISES STATISTIC

For an ordered sample  $x_1 \leq x_2 \leq \cdots \leq x_n$  the empirical distribution function is defined as

$$F_{n}(x) = \begin{cases} 0 & x < x_{1} \\ \frac{k}{n} & \text{for } x_{k} \le x < x_{k+1} \\ 1 & x_{n} \le x \end{cases}$$

The Cramér-von Mises statistic may be written

$$\begin{split} n & \int\limits_{-\infty}^{\infty} \left[ F_n(x) - F(x) \right]^2 \mathrm{d}F(x) \\ &= n \left\{ \int\limits_{-\infty}^{x_1} \left[ F_n(x) - F(x) \right]^2 \mathrm{d}F(x) \right. \\ &+ \sum\limits_{k=1}^{n-1} \int\limits_{x_k}^{x_{k-1}} \left[ F_n(x) - F(x) \right]^2 \mathrm{d}F(x) \right. \\ &+ \left. \int\limits_{x_n}^{\infty} \left[ F_n(x) - F(x) \right]^2 \mathrm{d}F(x) \right\} \end{split}$$

Let 
$$F(x) = u$$
  
 $dF(x) = du$   
 $F(x_i) = u_i$ 

Then

$$\begin{split} &n\int\limits_{-\infty}^{\infty}\left[F_{n}(x)-F(x)\right]^{2}dF(x)\\ &=n\left\{\int_{0}^{u_{1}}\left[0-u\right]^{2}du+\sum_{k=1}^{n-1}\int_{u_{k}}^{u_{k+1}}\left[\frac{k}{n}-u\right]^{2}du+\int_{u_{k}}^{1}\left[1-u\right]^{2}du\right\}\\ &=n\left\{\frac{1}{3}u_{1}^{3}+\sum_{k=1}^{n-1}\left[\frac{k^{2}}{n^{2}}u-\frac{k}{n}u^{2}+\frac{1}{3}u^{3}\right]_{u_{k}}^{u_{k+1}}+\left[\frac{1}{3}-u_{n}+u_{n}^{2}-\frac{1}{3}u_{n}^{3}\right]\right\}\\ &=n\left\{\frac{1}{3}u_{1}^{3}+\frac{1}{n^{2}}\sum_{k=1}^{n-1}k^{2}(u_{k+1}-u_{k})-\frac{1}{n}\sum_{k=1}^{n-1}k\left(u_{k+1}^{2}-u_{k}^{2}\right)\right.\\ &+\left.\frac{1}{3}\sum_{k=1}^{n-1}\left(u_{k+1}^{3}-u_{k}^{3}\right)+\left[\frac{1}{3}-u_{n}+u_{n}^{2}-\frac{1}{3}u_{n}^{3}\right]\right\}\\ &=n\left\{\frac{1}{3}u_{1}^{3}+\frac{1}{n^{2}}\left[\sum_{k=1}^{n-1}\left(-\left(2k-1\right)u_{k}\right)+\left(n-1\right)^{2}u_{n}\right]\right.\\ &-\left.\frac{1}{n}\left[-\sum_{k=1}^{n-1}u_{k}^{2}+\left(n-1\right)u_{n}^{2}\right]\right.\\ &+\left.\frac{1}{3}u_{n}^{3}-\frac{1}{3}u_{1}^{3}+\left[\frac{1}{3}-u_{n}+u_{n}^{2}-\frac{1}{3}u_{n}^{3}\right]\right\}\\ &=n\left\{\sum_{k=1}^{n-1}\left[\frac{1}{n}u_{k}^{2}-\left(\frac{2k-1}{n}\right)u_{k}\right]+\frac{1}{3}+\frac{1}{n}u_{n}^{2}-\left(\frac{2n-1}{n^{2}}\right)u_{n}\right\}\\ &=\sum_{n=1}^{n}\left[u_{k}^{2}-\left(\frac{2k-1}{n}\right)u_{k}\right]+\frac{n}{3}\end{split}$$

Completing the square,

$$\begin{split} &= \sum_{k=1}^{n} \left[ u_k - \left( \frac{2k-1}{2n} \right) \right]^2 + \frac{n}{3} - \sum_{k=1}^{n} \left( \frac{2k-1}{2n} \right)^2 \\ &= \sum_{k=1}^{n} \left[ u_k - \left( \frac{2k-1}{2n} \right) \right]^2 + \frac{n}{3} - \frac{1}{4n^2} \left( 1^2 + 3^2 + \dots + (2n-1)^2 \right) \\ &= \sum_{k=1}^{n} \left[ u_k - \left( \frac{2k-1}{2n} \right) \right]^2 + \frac{n}{3} - \frac{1}{4n^2} \left( \frac{(n)(4n^2 - 1)}{3} \right) \\ &= \sum_{k=1}^{n} \left[ u_k - \left( \frac{2k-1}{2n} \right) \right]^2 + \frac{1}{12n} \quad . \end{split}$$

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